Supplemental Material

Risk of Incident Diabetes in Relation to Long-term Exposure to Fine Particulate Matter in Ontario, Canada

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Comorbidity Ascertainment

- Ontario Hypertension Database
 - Data source: hospital discharge abstracts from Canadian Institute of Health Information (CIHI) (including same day surgery) database, physician service claims from Ontario Health Insurance Plan (OHIP) database
 - o Time period covered: 1988 onward
 - Case definition: one hospital admission with a hypertension diagnosis, or an OHIP claim with a hypertension diagnosis followed within two years by either an OHIP claim or a hospital admission with a hypertension diagnosis (ICD-9: 401-405; ICD-10: I10-I13, I15).
 - Sensitivity=72%, specificity=95%, positive predictive value=87%, and negative predictive value=88% (Tu et al. 2008).
- Ontario Asthma Database
 - Data source: hospital discharge abstracts from CIHI database, physician service claims from OHIP database
 - o Time period covered: 1991 onward
 - Case definition: one hospital admission with an asthma diagnosis or two OHIP claims with asthma diagnosis within a two-year period (ICD-9: 493; ICD-10: J45, J46).
 - Sensitivity=84% and specificity=76% in adults who are 18 years of age and older (Gershon et al. 2009a).
- Ontario Congestive Heart Failure (CHF) Database
 - Data source: CIHI discharge abstract database, physician service claims from the OHIP database, emergency department records from National Ambulatory Care Reporting System (NACRS)
 - o Time period covered: 1991 onward
 - Case definition: one hospital admission with a CHF diagnosis or an OHIP claim/
 emergency department record with a CHF diagnosis followed within two years by either

- a second OHIP claim/NACRS record or a hospital admission with a CHF diagnosis (ICD-9: 428; ICD-10: I500, I501, I509).
- o Sensitivity=85% and specificity=97% (Yeung et al. 2012).
- Ontario Chronic Obstructive Pulmonary Disease (COPD) Database
 - Data source: CIHI discharge abstract database, emergency department records from NACRS
 - o Time period covered: 1991 onward
 - Case definition: one or more ambulatory claims and/or one or more hospitalizations for COPD (ICD-9: 491, 492, 496; ICD-10: J41, J42, J43, J44).
 - o Sensitivity=85% and specificity=79% (Gershon et al. 2009b).
- Ontario Myocardial Infarction Database (OMID)
 - o Data source: CIHI discharge abstract database
 - o Time period covered: 1988 onward
 - Case definition: all patients with a most responsible diagnosis with ICD-9 410 or ICD-10 I21. Exclusion criteria are those who were not Ontario residents, who had a MI as a complication after admission to hospital, who were discharged with a total length of stay < 3 days, who were readmitted to hospital with a MI in the past year and those transferred from another acute care institution, or who were initially admitted to a noncardiac surgical service (Tu et al. 1999).</p>
 - o Sensitivity=89% and specificity=93% (Austin et al. 2002).

Assess Long-term Stability of the Satellite-based six-year Mean Concentrations of PM_{2.5}

We verified long-term stability in the spatial patterns of six-year average concentrations of PM_{2.5} over study period. In doing this, we compiled historical data on the monitoring of PM_{2.5} from Environment Canada's National Air Pollution Surveillance (NAPS) network (Environment Canada 2010). We excluded fixed-site monitors that were located outside Ontario and that were operated for less than half of the study period, leaving sufficient data to derive annual mean

concentrations for six cities in Ontario. These six cities are Toronto, Hamilton, Windsor, Ottawa, Simcoe, and St. Petre.

Using the monitoring data, we estimated for each city long-term average concentrations of PM_{2.5} over the entire study period. We compared the long-term averages of PM_{2.5} with the satellite-based six-year mean concentrations averaged among all study subjects in each city. Mean concentrations of PM_{2.5} between the two periods were reasonably well correlated (Pearson's correlation coefficient r = 0.77).

Using the annual mean concentrations of $PM_{2.5}$ from the six cities, we further estimated the total variance of $PM_{2.5}$ across the six cities and throughout the study period between 1996 and 2010. In addition, we estimated the variance of $PM_{2.5}$ that was due to temporal variability from 1996 to 2010. This was done by calculating mean exposure averaged across the six cities for each year and then estimating the variance of the annual averages over time. The total variance was 6.70 $(\mu g/m^3)^2$ while the temporal variance was 2.25 $(\mu g/m^3)^2$. Thus, 67% of the total variation in the concentrations of $PM_{2.5}$ among the six cities between 1996 and 2010 is associated with spatial variability and only 33% with variation over time. This result suggests that variability in the concentrations of $PM_{2.5}$ in Ontario is primarily spatial in nature and not temporal. This finding is reinforced by the fact that the rank ordering of the six Ontario cities by relative levels of $PM_{2.5}$ during the study period remained nearly constant (Figure S1).

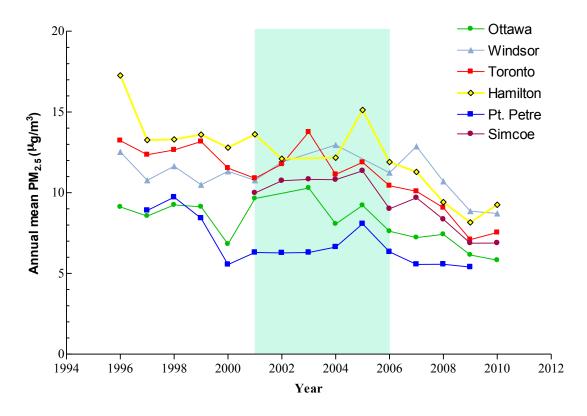
The representativeness of shorter-term PM_{2.5} measurements for longer-term exposure has been reported in several previous studies (Jerrett et al. 2005; Miller et al. 2007; Pope et al. 2002). For example, in the American Cancer Society Cancer Prevention Study II (ACS study) that was conducted in Los Angels, California, Jerrett *et al.* (2005) assessed the relationship between PM_{2.5}

measured in 1980 and those in 1999-2000 at 51 fixed-site monitors. They found strong correlation in $PM_{2.5}$ measurements between the two periods (the coefficient of determination, R^2 =61% or r=0.78), indicating that areas with higher particle concentrations in earlier periods were likely to retain their spatial ranking. Long-term stability in the spatial patterns of $PM_{2.5}$ has also been demonstrated in another ACS study that included entire ACS cohort from 116 metropolitan areas in the U.S.A (Pope et al. 2002) and in the Women's Health Initiative Study that comprised subjects from 36 metropolitan areas in the U.S.A (Miller et al. 2007). We therefore expect that the spatial contrast in $PM_{2.5}$ over 2001-2006 provided reasonable estimates of longer-term spatial exposure to $PM_{2.5}$ in Ontario.

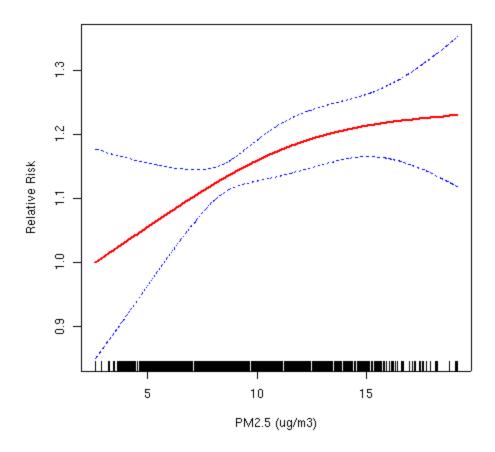
Comparison of Spatial Resolution for Different Datasets in the Study

Spatial resolution for different datasets used in our study is described as follows: Postal codes (a total of 269,676 in Ontario) > Census tract (2,136 in Ontario) > 10km by 10km grids in the $PM_{2.5}$ exposure surface (1,198 grids in Ontario) > Census division (50 in Ontario) > Ontario local health integration networks or LIHN (a total of 14 in Ontario).

These datasets were created by different organizations for different purposes; as a result, their areas may overlap. For example, LIHNs may overlap with census divisions.



Supplemental Material, Figure S1. Trends in annual average concentrations of $PM_{2.5}$ (in $\mu g/m^3$) in six cities in Ontario, Canada between 1996 and 2010. Data were obtained from Environment Canada's National Air Pollution Surveillance (NAPS) network. Fixed-site monitors operated for less than half of the entire study period (<8 years) were excluded. The shaded area denotes the time period during which satellite-based surface measurements of $PM_{2.5}$ were available.



Supplemental Material, Figure S2. Concentration-response relationship between the concentration of PM_{2.5} and incident diabetes among the cohort, depicted using a natural cubic spline function with 2 degrees of freedom. The hazard ratios were estimated by comparing to 2.6μg/m³. The Cox model stratified by age, survey year and region, and adjusted for sex, marital status, education, household income, BMI, physical activity, smoking, alcohol consumption, diet, race, hypertension, urban residency, neighborhood-level unemployment rate, education, household income, and COPD, asthma, congestive heart failure, and acute myocardial infarction.

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